

SLAM Demo

EDAA — G06

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Simultaneous Location And Mapping

- **Goal** – map an environment navigated by and autonomous vehicle, while simultaneously locating it in the map;
- **Challenges:**
 - Move from 2D space to 3D space - Raycast volume estimation
 - Optimize parallel raycasts - Hash Sets
 - Improve obstacle and edge detection - Canny filter
- **Datasets** – the group will have access to sonar data measured by CRAS.

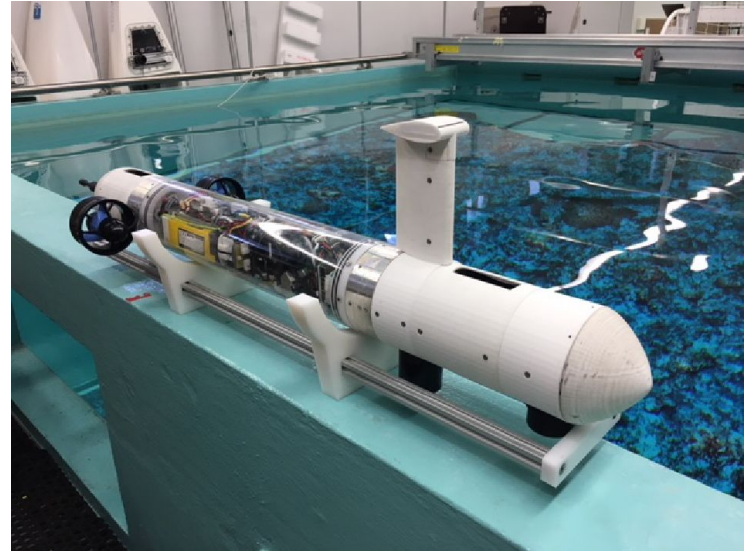
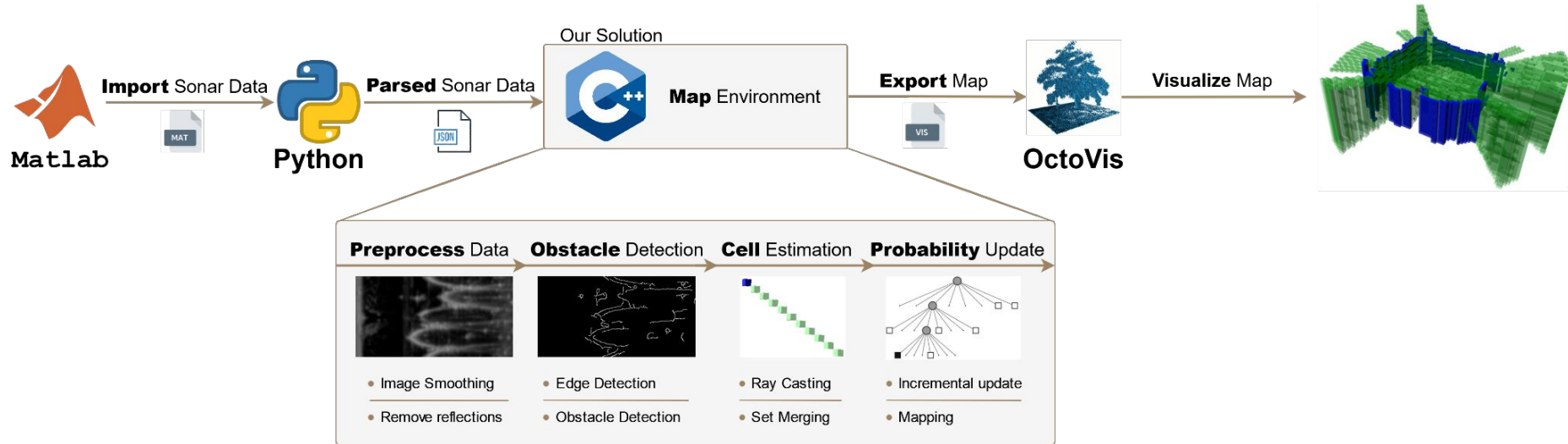


Fig 1. UAV used to collect the datasets.

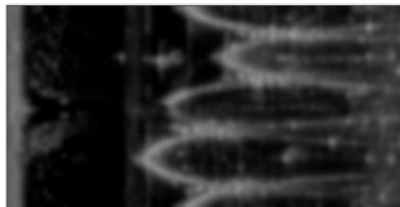
Pipeline

Pipeline



Our Solution

Preprocess Data



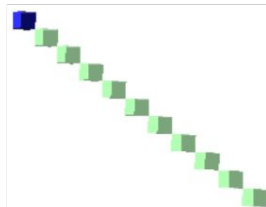
- Image Smoothing
- Remove reflections

Obstacle Detection



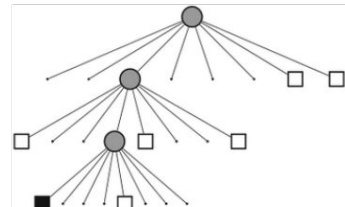
- Edge Detection
- Obstacle Detection

Cell Estimation



- Ray Casting
- Set Merging

Probability Update

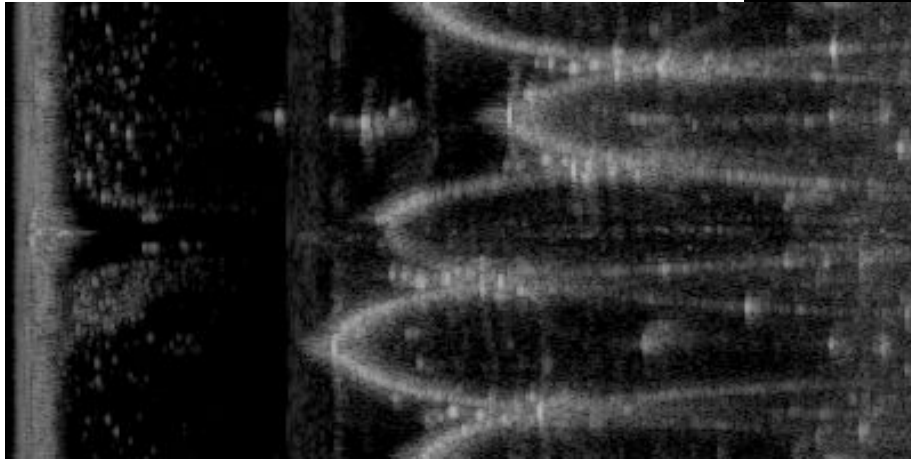


- Incremental update
- Mapping

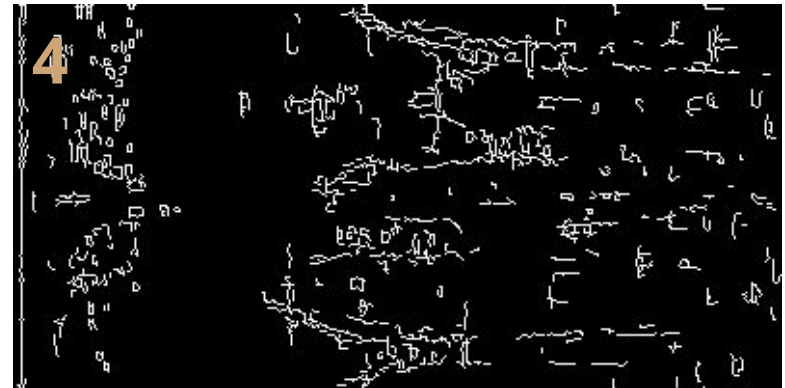
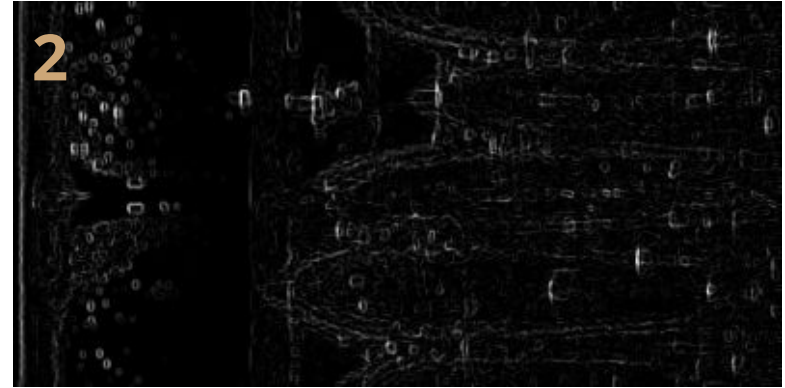
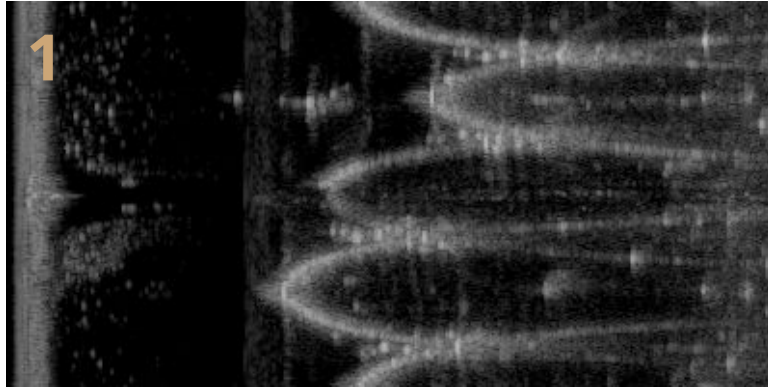
Obstacle Detection

- Canny Thresholding

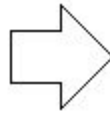
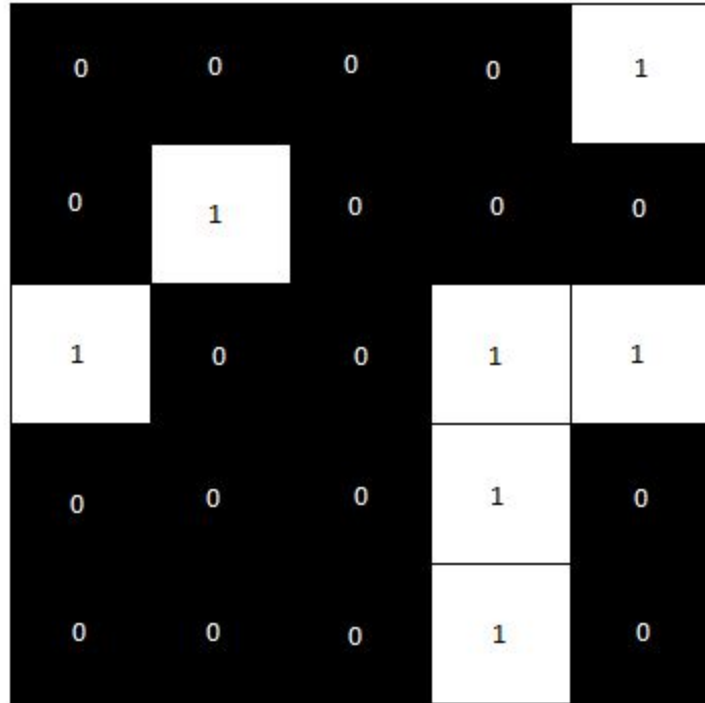
Obstacle Detection - Canny threshold



Canny Intermediate Results

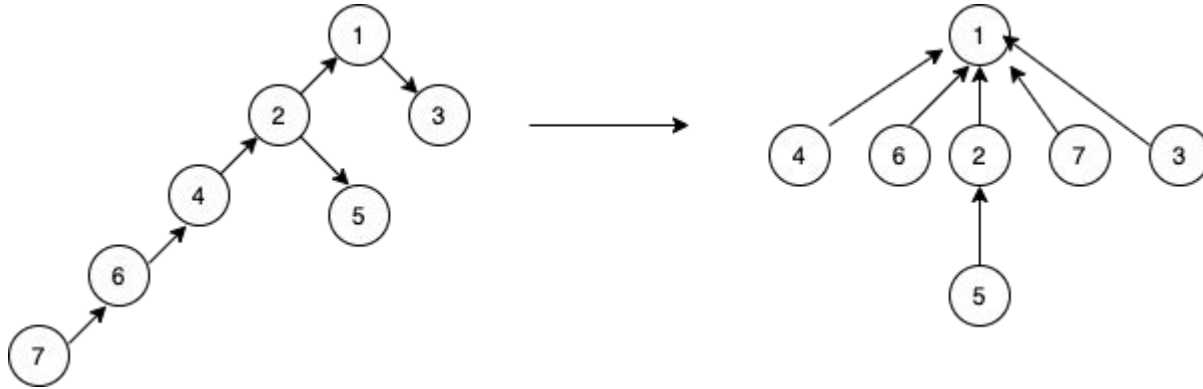


Hysteresis and Blob Analysis



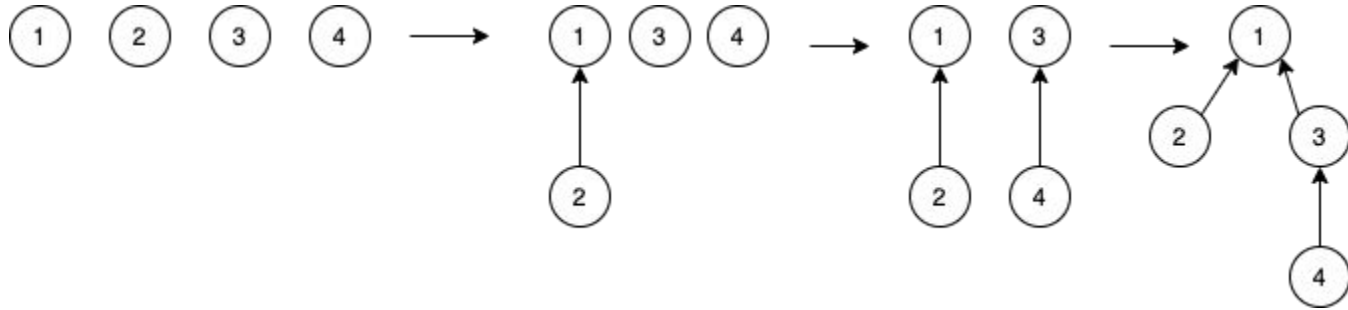
Disjoint Sets

- Elements partitioned into a number of disjoint sets
- Using Path Compression

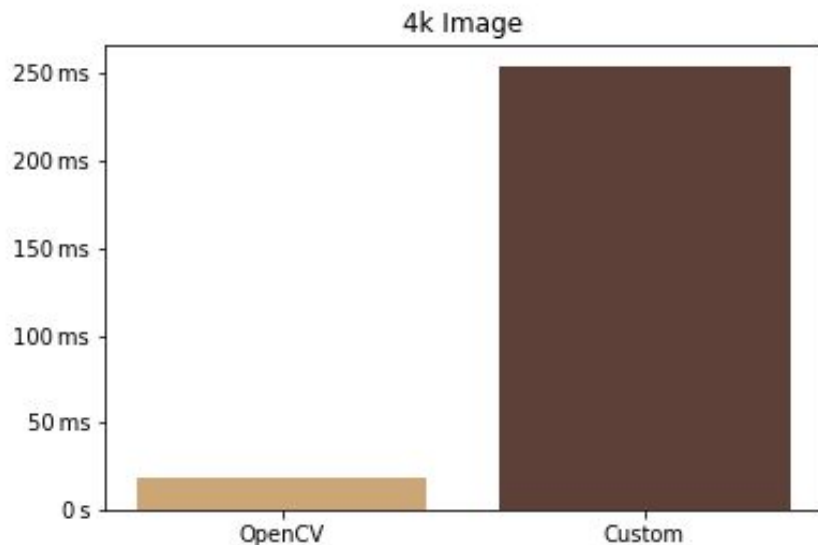
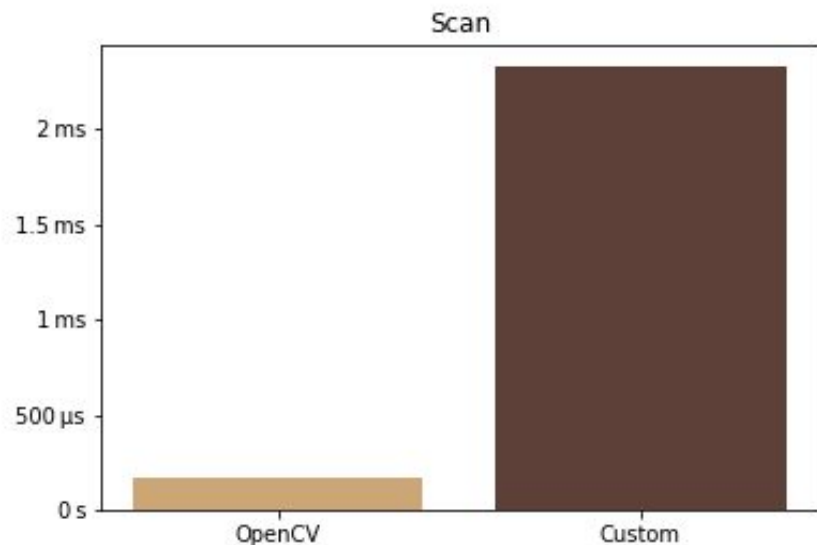


Disjoint Sets - Applied to our Blob Analysis

- Keep the smallest element at the root of each set
- ✖ Cannot implement Union by Rank



Performance



Cell Estimation

- Ray Casting
- Set Merging

Raycasting

- Bresenham's line algorithm:
 - Travels in all axis at once
 - Handles diagonal transitions very well
 - Very accurate
- Several beams are sent at once
 - Overlapping cells that need to be merged

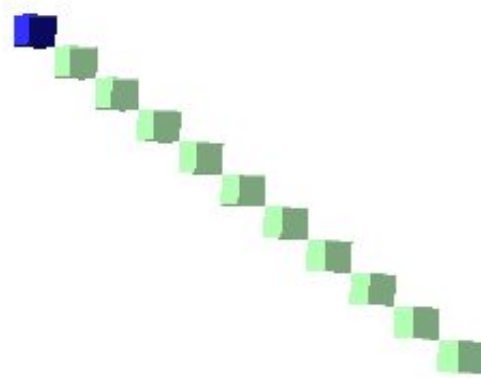


Fig 2. Bresenham's Line Algorithm

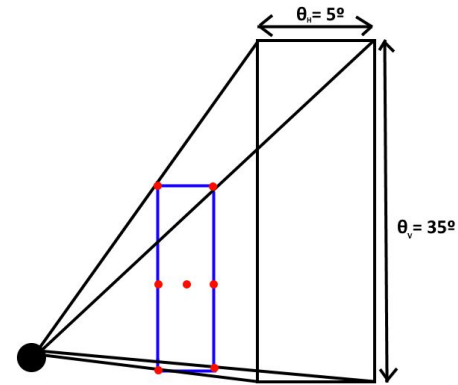


Fig 3. Volume estimation with raycasts

Sets using Hash Maps

- Used to join covered cells from parallel raycasts
- Set merge operation used 60% of ray cast's execution time;

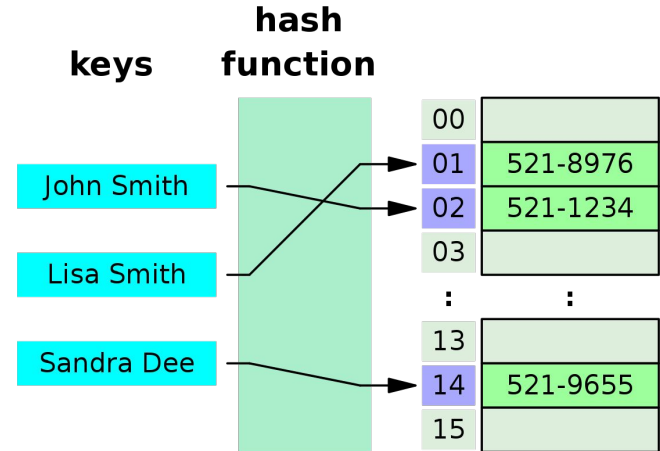


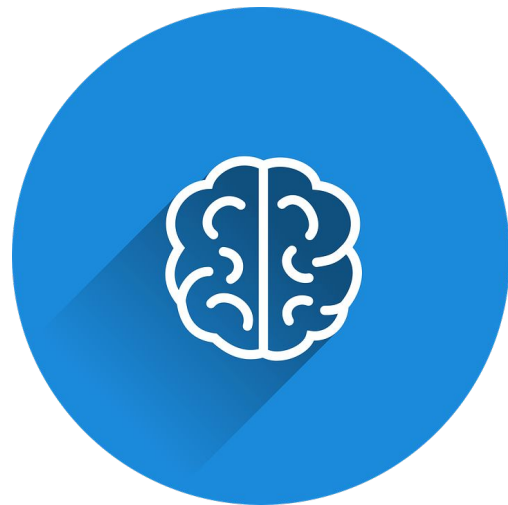
Fig 4. Hash Map Representation

Set/Hash table

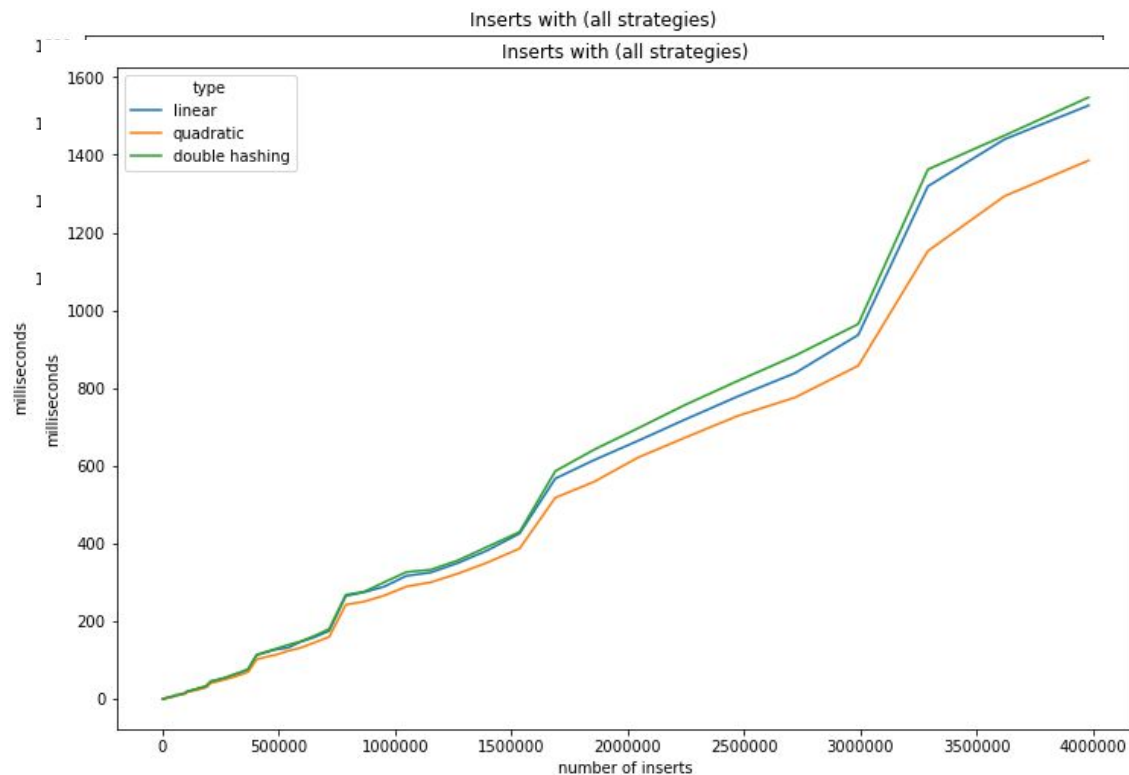
- Implementing sets (based on Hash tables)
- Focus on improving performance on our use case

Rationale – Opportunities for improvements

- Our use case needs unordered sets as containers for very small objects:
 - Objects are 48-bit < 64-bit pointers.
- C++ sets use buckets:
 - Increases memory usage.
- C++ sets aren't fast for our use case.

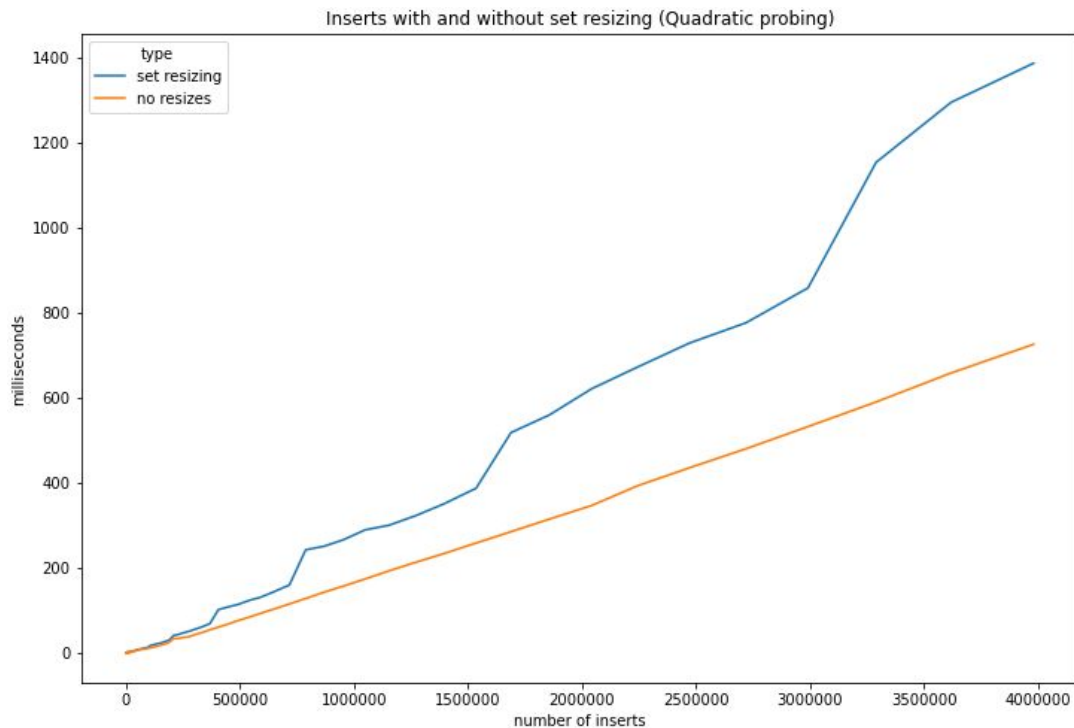


Insert – Comparison of all strategies



- Insertion of randomly generated objects;
- Performance is similar between all methods;
- **Quadratic probing** is consistently the best.

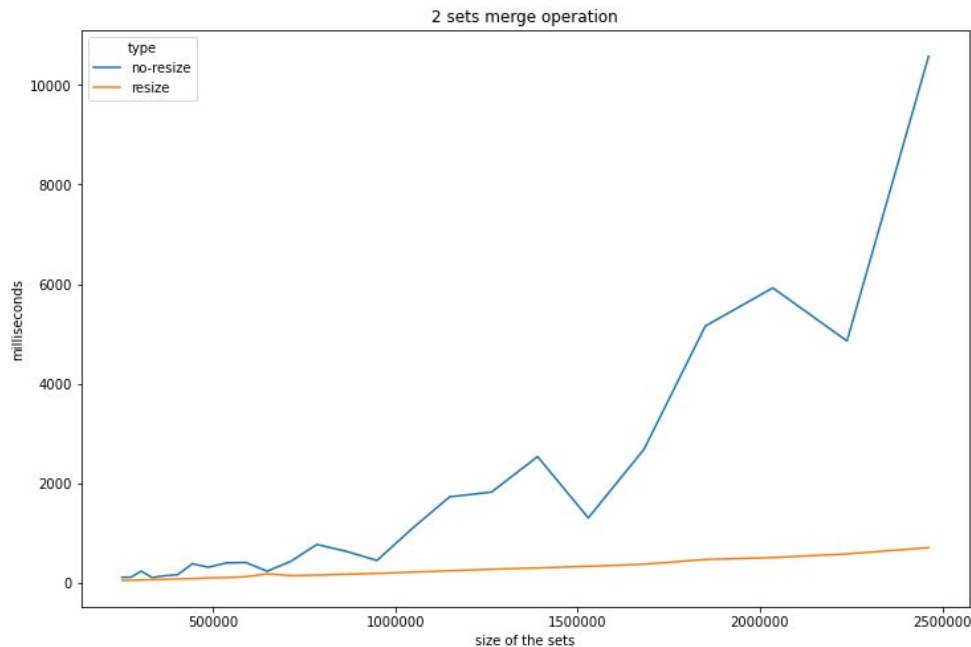
Insert — Impact of set resizing



- In the previous examples, the sets were resized multiple times;
- This also affects the comparisons:
 - Smaller sets suffer fewer resizes.
- Resize is triggered by a threshold of how full the set is (**75%**);

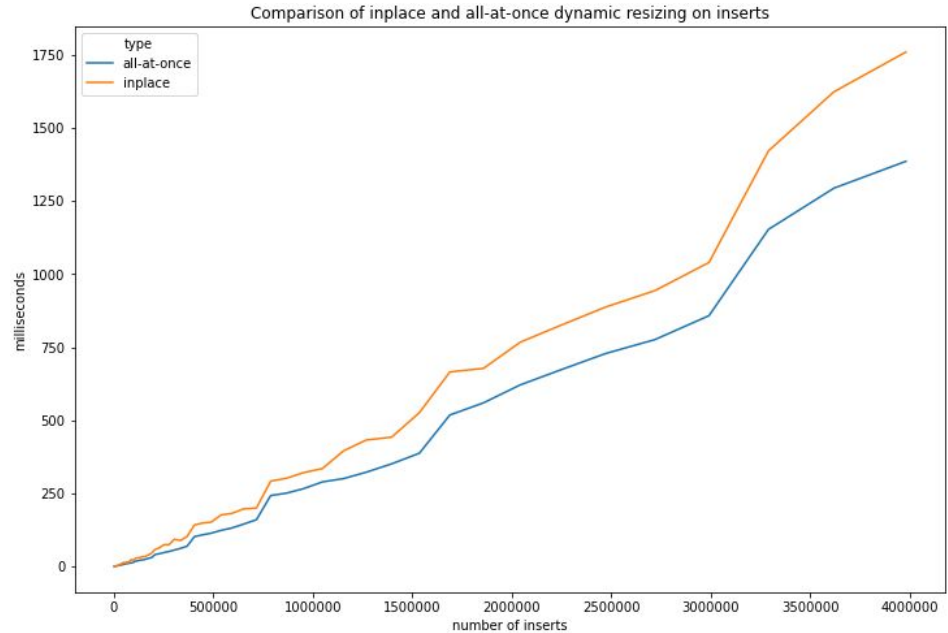
Merge — Performance and impact of resize

- In our project, we only need **in-place merges**;
- The **merge** operation consists of iterating over a set and inserting each object into the second one;
- By doing a **pessimistic assumption**, we can greatly increase the performance of the operation:
 - Assume that each element in the second set is new (not already part of the set);
 - Check that the **load factor is still ok** in that case (no resize needed).



In-place dynamic resizing – comparison

- We implemented a way for sets to be dynamically resized in-place:
 - Less memory overhead;
 - Less time performance.
- Algorithm:
 - Grow the container (double size);
 - Re-evaluate all entries in first half;
 - If they fall on second half, place them there;
 - Otherwise, save them in a buffer for later;
 - Clear first half and insert the buffer elements



Real data test

- Tests on real data:
 - Ray-casts on a plane point cloud dataset;
 - Only considering the turbines;
 - About 1000 ray-casts covering 18000 cells each.
- Ray-casts are calculated 4 at a time in parallel:
 - On a computer with 4 CPU cores.
- Our set implementation **improved performance almost 2-fold.**

| | C++ sets | Our sets | In-place resize |
|---------------|----------|----------|--------------------|
| Time taken | ≈ 23 s | ≈ 13 s | ≈ 10 s |

Probability Update

- Incremental Update

Incremental Probability Update

- Build map probabilities on top of previous sweep measurements

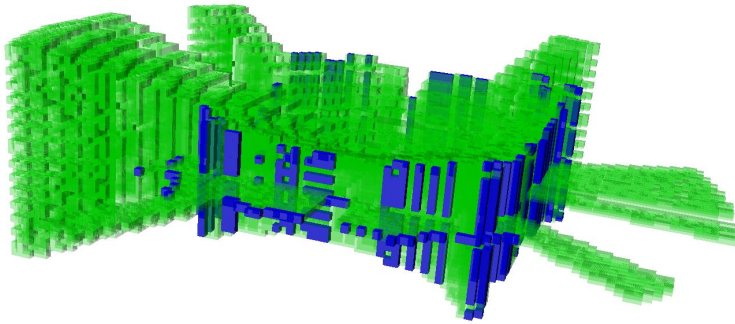


Fig 5. Map after first sweep

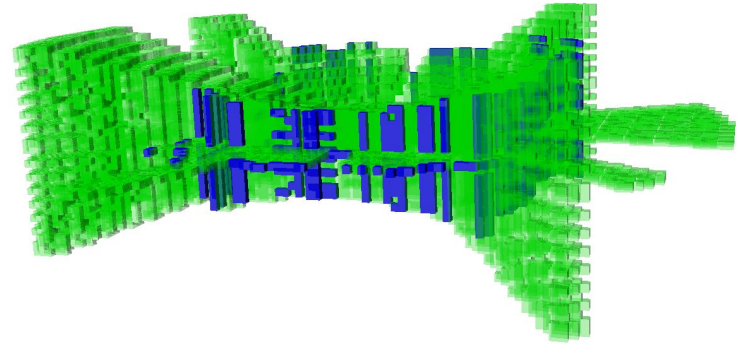
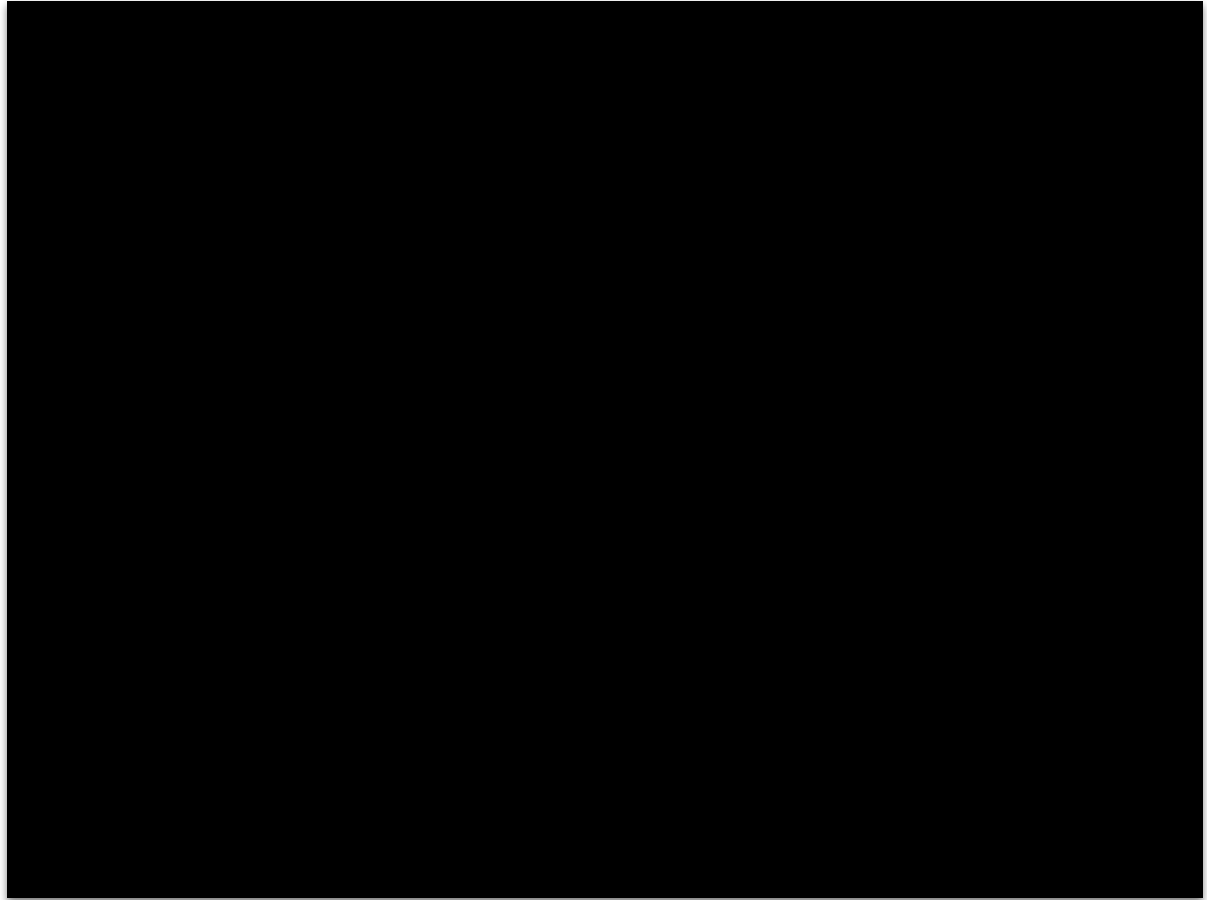


Fig 6. Map after 27 sweeps

Demo

- Results showcasing
- Demo video

Results



Future work

- Aspects to explore
- Things to improve

Future work

- Explore **fast modulo** operation (as defined by Boost);
- Try different set collision resolution techniques;
- Use OpenCV's Canny algorithm;
- Explore different SLAM datasets;
- Deal with the discontinuities found in the walls of the tank;
- Explore alternatives to raycasts;
- Explore alternative probability distributions.

Slide deck source

Slide deck